

In the Claims:

Please cancel claims 5, 9-12 and 20-24, without prejudice; amend claims 1, 6-8 and 13-16; and add new claims 25-50 as follows:

- 1 1. (Amended) A method of determining spatial target probability  
2 comprising the steps of:  
3 acquiring at least two inputs from a location in a desired environment;  
4 applying said inputs to a plurality of model units in a map corresponding to a  
5 plurality of locations in said environment;  
6 approximating a posterior probability of a first target at each of said model  
7 units based on said at least two inputs;  
8 finding a model unit from said plurality of model units with a highest posterior  
9 probability;  
10 choosing a location in said environment corresponding to said model unit with  
11 said highest posterior probability as a location of a next target.

Please cancel claim 5 without prejudice.

- 1 6. (Amended) The method as defined in claim 4, wherein said posterior  
2 probability is approximated using a sigmoid curve function.

- 1 7. (Amended) The method as defined in claim 4, wherein said posterior  
2 probability is approximated using a linear function.

4 applying at least one input associated with said training target to the model  
5 neural network to obtain said actual responses of the model neural network to said training  
6 target;  
7 generating said desired responses of the model neural network to said training  
8 target;  
9 finding differences between said actual and desired responses; and  
10 using back-propagation to reduce said differences between said actual and  
11 desired responses.

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1 15. (Amended) An apparatus for automatically tracking a target in a desired  
2 environment, said system comprising:  
3 at least one first sensor for receiving sensory inputs from the target;  
4 a controller, based on said sensory inputs, for locating the target in the  
5 environment using a program modeling a neural network of a brain; and  
6 at least one directional second sensor for turning to a location in the  
7 environment where the target has been located by said controller,  
8 wherein said model of said neural network includes a map having a plurality of  
9 model units corresponding to a plurality of locations in the environment for receiving  
10 information from said sensory inputs associated with the target located in the environment  
11 through a plurality of input units and connections between said input units and said model  
12 units.

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- 1 16. (Amended) The apparatus as defined in claim 15 wherein the target is  
2 located by approximating a posterior probability of the target given said sensory inputs.
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Please cancel claims 20-24 without prejudice.

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- 1 25. (New) The method as defined in claim 13, wherein the plurality of  
2 output units of the model neural network represent model units in a map corresponding to a  
3 plurality of locations in said desired environment.

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- 1 26. (New) The method as defined in claim 14, wherein said step of using  
2 back-propagation includes iteratively adjusting weights associated with the hidden units.

- 1 27. (New) The method as defined in claim 13, wherein said predetermined  
2 inputs and said at least one input associated with said first target are sensory inputs.

- 1 28. (New) The method as defined in claim 27, wherein said sensory inputs  
2 include audio and video inputs.

- 1 29. (New) A method of determining spatial target probability using an  
2 unsupervised adaptive algorithm in a model of a neural network, said method comprising the  
3 steps of:

4 organizing a map into a plurality of model units corresponding to a plurality of  
5 locations in a desired environment for receiving information from sensory inputs associated  
6 with a target located in said environment through a plurality of input units and connections  
7 between said input units and said model units;

8 adjusting said map to increase an amount of said information from said sensory  
9 inputs that are transmitted to said map using an unsupervised learning mechanism; and

10 modulating a strength of said sensory inputs associated with said target based  
11 on a correlation between activities of said map and predefined modulation units, and on anti-  
12 correlation between said predefined modulation units and said sensory inputs associated with  
13 said target.

1 30. (New) The method as defined in claim 29, wherein a Kohonen  
2 mechanism is used in said step of adjusting said map.

1 31. (New) The method as defined in claim 30 wherein said Kohonen  
2 mechanism adjusts weights associated with said connections between said input units and  
3 said model units such that each of said model units become specialized for receiving  
4 information indicating a predetermined location in said environment.

1 32. (New) The method as defined in claim 31 wherein said model units  
2 further become specialized for receiving a predetermined modality of said sensory inputs  
3 associated with said target.

1                    33. (New) The method as defined in claim 32 wherein said modality  
2 includes at least audio and video inputs.

1                    34. (New) The method as defined in claim 32 wherein said modulation  
2 units are predefined according to a modality of said sensory inputs associated with said  
3 target.

1                    35. (New) The method as defined in claim 34 wherein said modulation  
2 units modulate said strength of said sensory inputs by multiplying weights associated with  
3 said sensory inputs.

1                    36. (New) The method as defined in claim 34 wherein said modality  
2 includes at least audio and video inputs.

1                    37. (New) The method as defined in claim 34 wherein at least one of said  
2 modulation units predefined by a first modality of said sensory inputs becomes active when  
3 said map receives information through at least said first modality from said sensory inputs,  
4 said at least one of said modulation units decreases modulation of said sensory inputs having  
5 first modality and increases modulation of said sensory inputs having modality other than  
6 said first modality.

1 38. (New) The apparatus as defined in claim 15 wherein said at least one  
2 first sensor includes at least one audio and at least one video sensor.

1 39. (New) The apparatus as defined in claim 38 wherein said sensory  
2 inputs are audio and video signals.

1 40. (New) The apparatus as defined in claim 15 wherein said at least one  
2 directional second sensor includes at least one of an audio and a video sensor.

1 41. (New) The apparatus as defined in claim 15 wherein the target is  
2 located by a supervised learning algorithm in which,

3 said model neural network is trained to reduce an error between an actual  
4 response and a desired response of said model neural network to predetermined inputs from a  
5 known location in the environment;

6 sensory inputs associated with the target located in the environment is applied  
7 to said plurality of inputs of said model neural network;

8 the model units with a highest desired value is found; and

9 a location in the environment corresponding to said model unit with said  
10 highest desired value is chosen as a location of a next target.

1 42. (New) The method as defined in claim 41, wherein said training of said  
2 model neural network includes,

3 positioning a training target at a random location in the predefined  
4 environment;  
5 applying sensory inputs associated with said training target to the model neural  
6 network to obtain said actual responses of the model neural network to said training target;  
7 generating said desired responses of the model neural network to said training  
8 target;  
9 finding differences between said actual and desired responses; and  
10 using back-propagation to reduce said differences between said actual and  
11 desired responses.

1 43. (New) The apparatus as defined in claim 15 wherein the target is  
2 located by an unsupervised adaptive algorithm in which,  
3 said map is adjusted using a Kohonen mechanism to increase an amount of  
4 information from said sensory inputs that are transmitted to said map; and  
5 a strength of said sensory inputs associated with the target is modulated based  
6 on a correlation between activities of said map and predefined modulation units, and on anti-  
7 correlation between said predefined modulation units and said sensory inputs associated with  
8 said target.

1 44. (New) The apparatus as defined in claim 43 wherein said Kohonen  
2 mechanism adjusts weights associated with said connections between said input units and

3 said model units such that each of said model units become specialized for receiving  
4 information indicating a predetermined location in said environment.

1 45. (New) The apparatus as defined in claim 44 wherein said model units  
2 further become specialized for receiving a predetermined modality of said sensory inputs  
3 associated with said target.

1 46. (New) The apparatus as defined in claim 45 wherein said modality  
2 includes at least audio and video inputs.

1 47. (New) The apparatus as defined in claim 45 wherein said modulation  
2 units are predefined according to a modality of said sensory inputs associated with said  
3 target.

1 48. (New) The apparatus as defined in claim 47 wherein said modulation  
2 units modulate said strength of said sensory inputs by multiplying weights associated with  
3 said sensory inputs.

1 49. (New) The apparatus as defined in claim 47 wherein said modality  
2 includes at least audio and video inputs.



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1           50.   (New) The apparatus as defined in claim 47 wherein at least one of said  
2 modulation units predefined by a first modality of said sensory inputs becomes active when  
3 said map receives information through at least said first modality from said sensory inputs,  
4 said at least one of said modulation units decreases modulation of said sensory inputs having  
5 first modality and increases modulation of said sensory inputs having modality other than  
6 said first modality.

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